CS 396: Online Markets

Lecture 12: Revenue Maximization

Last Time:

- learning to bid (cont)
- partial feedback
- equilibrium of no-regret learning (coarse correlated equilibrium)

Today:

- equilibrium of no-regret learning (coarse correlated equilibrium)
- revenue of auctions

Exercise: Optimal Pricing

Setup:

- you have one item to sell.
- buyer with value from exponential distribution
- exponential distribution cdf $F(z) = 1 e^{-z}$

Questions:

• what price should you offer to maximize your expected revenue?

Auction Revenue

Recall:

- two bidders, values U[0, 1]
- second-price auction
 - bids = values

$$-\mathbf{E}[\text{revenue}] = \mathbf{E}[\mathbf{b}_{(2)}] = \mathbf{E}[\mathbf{v}_{(2)}] = 1/3$$

- first-price auction
 - bids = half of values
 - $-\mathbf{E}[\text{revenue}] = \mathbf{E}[\mathsf{b}_{(1)}] = \mathbf{E}[\mathsf{v}_{(1)}/2] = 1/3$
- same revenue

Def: second-price auction with reserve \hat{v}

- highest bidder with bid at least \hat{v} wins

- pays maximum of second-highest bid and \hat{v}

Same thing as if seller entered a bid of $\hat{v}.$

Example:

- two bidders, uniform values,
- reserve $\hat{\mathbf{v}} = 1/2$.
- revenue calculation:

| Case | Probability | expected revenue |
|--|-------------|------------------|
| $\overline{v_{(2)} < v_{(1)} < \hat{v}}$ | 1/4 | 0 |
| $v_{(2)} < \hat{v} < v_{(1)}$ | 1/2 | 1/2 |
| $\hat{v} < v_{(2)} < v_{(1)}$ | 1/4 | 2/3 |

• total: $1/4 \times 0 + 1/2 \times 1/2 + 1/4 \times 2/3 = 5/12 > 1/3$

Note: by sometimes withholding the item, the seller can make more revenue.

Q: is there an auction with higher expected revenue?

A: no.

- Roger Myerson (1981) while at Kellogg
- 2007 Nobel prize in economics
- all you need to know is integration by parts!

Revenue Optimal Pricing

"optimal price to post a single buyer"

- buyer value $v \sim F$
- revenue from price $\hat{\mathbf{v}}$: $\hat{\mathbf{v}}\mathbf{Pr}[\mathbf{v} \geq \hat{\mathbf{v}}] = \hat{\mathbf{v}}(1 F(\hat{\mathbf{v}}))$
- to optimize $\frac{d}{d\hat{\mathbf{v}}}[\hat{\mathbf{v}}(1-F(\hat{\mathbf{v}}))] = 1 F(\hat{\mathbf{v}}) \hat{\mathbf{v}}f(\hat{\mathbf{v}}) = 0$
- note: tradeoff $\hat{\mathbf{v}}$ with probability of sale $1 F(\hat{\mathbf{v}})$.

Example:

- $v \sim U[0, 1]$, i.e., F(z) = z
- $\frac{d}{d\hat{x}}[\hat{v}(1-\hat{v})] = 1 2\hat{v} = 0$
- $\Rightarrow \hat{\mathbf{v}} = 1/2.$
- expected revenue is $1/2 \times (1 F(1/2)) = 1/4$.

Exercise: Pricing Lotteries

Setup:

- buyer with value U[0,1]
- menu of options:
 - (a) price of 0: receive nothing
 - (b) price of 1/6: receive item with probability 1/2
 - (c) price of 1/2: receive item with probability 1

Questions:

- what value of buyer is indifferent between (a) and (b)?
- what value of buyer is indifferent between (b) and (c)?
- what is expected revenue when buyer buys preferred option?

Revenue Analysis of Lotteries

"e.g., in multi-bidder auction, outcome depends on other bids"

Def: allocation rule: $x(\mathbf{v}) = \mathbf{Pr}[\text{bidder with value } \mathbf{v} \text{ wins}]$

Def: posted pricing implementation of x:

- draw price \hat{v} from distirbution G(z) = x(z)
- $x(z) = \mathbf{Pr}[\text{bidder offered price } \hat{v} \leq v] = G(z)$

Lemma: revenue from x is $\int_0^\infty z(1 - F(z))x'(z)dz$.

Proof:

- density function for $\hat{\mathbf{v}}$ is $x'(\hat{\mathbf{v}})$.
- revenue if offer price $\hat{\mathbf{v}}$ is $\hat{\mathbf{v}}(1 F(\hat{\mathbf{v}}))$
- definition of expectation.

Thm: revenue from x is $\mathbf{E}\left[\left[\mathsf{v} - \frac{1-F(\mathsf{v})}{f(\mathsf{v})}\right] x(\mathsf{v})\right]$.

Recall: integration by parts

$$\int_{a}^{b} f(z) g'(z) dz = [f(z)g(z)]_{a}^{b} - \int_{a}^{b} f'(z) g(z) dz$$

Proof:

• integration by parts: $\int_0^\infty \mathsf{z}(1-F(\mathsf{z}))x'(\mathsf{z})d\mathsf{z}$ $= [z(1 - F(z))x(z)]_0^{\infty} - \int_0^{\infty} [1 - F(z) - z f(z)] x(z) dz$

- simplify with z(1 F(z)) = 0 at z = 0 and $z = \infty$
- factor out density to write as expectation: $\mathbf{E}_{\mathbf{v}\sim F}\left[\left[\mathbf{v}-\frac{1-F(\mathbf{v})}{f(\mathbf{v})}\right] x(\mathbf{v})\right]$

Recall: welfare is value of winner, expected welfare is $\mathbf{E}[v x(v)]$

Def:

- virtual value is φ(v) = v 1-F(v)/f(v)
 virtual welfare is E[φ(v) x(v)]

Conclusion: expected revenue = expected virtual welfare

Q: how to maximize virtual surplus?

A: allocate if $\varphi(\mathbf{v}) \geq 0$, i.e., if $\mathbf{v} \geq \varphi^{-1}(0)$

Corollary: optimal single-buyer mechanism posts price $\varphi^{-1}(0)$.

Conclusion: selling lotteries does not improve revenue.

Example:

•
$$\mathbf{v} \sim U[0, 1]$$

•
$$F(\mathbf{v}) = \mathbf{v}; f(\mathbf{v}) = 1$$

•
$$\varphi(\mathbf{v}) \equiv \mathbf{v} - \frac{1}{1} \equiv 2\mathbf{v} - 1$$

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• $\varphi^{-1}(0) = 1/2$

A: optimal mechanism

Q: what is virtual surplus? **A:** 1/4.

Q: what mechanism does this? **A:** post price of 1/2

Multi-bidder Mechanisms

Thm: in multi-bidder mechanisms, expected revenue equals expected virtual welfare.

Q: how to maximize virtual welfare in single-item auction?

A: allocate to bidder with highest positive virtual value.

e.g., bidder 1 wins if $\varphi_1(\mathsf{v}_1) \ge \max(\varphi_2(\mathsf{v}_2), 0)$, $\Rightarrow 1$ wins if $\mathsf{v}_1 \ge \max(\varphi_1^{-1}(\varphi_2(\mathsf{v}_2)), \varphi_1^{-1}(0))$

e.g., i.i.d. values (i.e., $\varphi_1 = \varphi_2 = \varphi$) $\Rightarrow 1$ wins if $\mathbf{v}_1 \ge \max(\mathbf{v}_2), \varphi^{-1}(0)$)

Q: what auction does this?

A: second-price auction with reserve $\hat{\mathbf{v}} = \varphi^{-1}(0)$

Cor: for i.i.d. buyers second-price auction with reserve $\varphi^{-1}(0)$ is revenue optimal

Example: two bidders, values U[0, 1]

- $\varphi^{-1}(0) = 1/2$
- (from example) 5/12 is optimal revenue.

Equilibrium of No-regret Learning

"outcomes of games under learning"

Def: coarse correlated equilibrium (CCE)

- mediator offers joint distributions of actions (r, c)
- players either
 - follow mediator
 - pick any fixed outcome
- CCE if best response is to follow mediator

Thm: play is no-regret iff emprical distribution of play is CCE.

- fix sequence $((r^0, c^0), \dots, (r^n, c^n))$
- **no-regret** for player R, for all actions r^* of R: $\sum_i R_{r^i,c^i} \ge \sum_i R_{r^*,c^i}$
- consider mediator:
 - pick *i* uniformly from round $\{1, \ldots, n\}$
 - recommend r^i to R and c^i to C.
- no-regret for R and C \Leftrightarrow mediator is CCE.